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GRIDS

Miriah Meyer University of Utah





administrivia . . .

-parallel coordinates assignment due tonight

-scalar data assignment out today

last time . . .

MAPS







The New york Times | ASIA

Map of the Damage From the Japanese Earthquake

An interactive map and photographs of places in Japan that were damaged by the March 11 earthquake and tsunami.



data as points

data : ordered/ quantitative

encoding : size

isopleth

map which overlays continuous data using a third encoding channel

Lines of Equal Magnetic Declination first contour map



Edmond Halley, 1701

choropleth

map in which areas are shaded, colored, or patterned relative to a data attribute value

Illiteracy in France

first choropleth map



Charles Dupin, 1826

cartogram

map in which areas are scaled and distorted relative to a data attribute value









equal-area preserves area

conformal



























-data sources

-data representation

-interpolation

DATA SOURCES

data sources

- Medical Imaging (MRI, CT, PET)
- Geographical information systems (GIS)
- Electron microscopy
- Meteorology and environmental sciences (satellites)
- Seismic data
- Crystallography
- High energy physics
- Astronomy (e.g. Hubble Space Telescope 100MB/day)

– Defense







GB

MB

THEORETICAL MEASUREMENTS

- Sciences

- Molecular dynamics
- Quantum chemistry
- Mathematics
- Molecular modeling
- Computational physics
- Meteorology
- Computational fluid mechanics (CFD)
- Engineering
 - Architectural walk-throughs
 - Structural mechanics
 - Car body design





MB

GB

GB



DATA REPRESENTATION

• Discrete representations

- objects we want to visualize are continuous
- but, data only given at discrete locations
- grids (meshes) consist of cells generated from data points

• Primitives in different dimensions

dimension	cell	mesh
0D	points	1 1• / \
1D	lines (edges)	polyline(–gon) 2D mesh 3D mesh
2D	triangles, quadrilaterals (rectangles)	
3D	tetrahedra, prisms, hexahedra	

- dimension of domain (the field)
- dimension of the data to visualize (the geometry)



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- Visualization of 1D, 2D, or 3D scalar fields
 - 1D scalar field: $\Omega \in R \to R$
 - 2D scalar field: $\Omega \in R^2 \to R$
 - 3D scalar field: $\Omega \in R^3 \to R$

 \rightarrow Volume visualization!

- Mapping to geometry
 - Function plots
 - Height fields
 - Isolines and isosurfaces
- Color coding
- Specific techniques for 3D data
 - Indirect volume visualization
 - Direct volume visualization
 - Slicing
- Visualization method depend heavily on dimensionality of domain









- -NIH project established in 1989 -male: 1994
 - -1,871 4mm slices
 - -15GB
- -female: 1995
 - -5,189 0.33mm slices
 - -40GB
- -MRI, CT, and color



-1,871 4mm slices

-15GB

-40GB

-female: 1995

-5,189 0.33m

The National Library of Medicine's

Visible Human Project (TM)

Human-Computer Interaction Lab Univ. of Maryland at College Park

-MRI, CT, and color










T. Fogal, J. Krueger. Size Matters - Revealing Small Scale Structures in Large Datasets, In IFMBE Proceedings, Vol. 25/13, Springer Berlin Heidelberg, pp. 41--44. 2009.

- Representation of scalar 3D data set $\Omega \in R^3 \to R$
- Analogy: pixel (picture element)



- Voxel (volume element), with two interpretations:
 - Values between grid points are resampled by interpolation



- Collection of voxels
- Uniform grid





Input Data

- Discrete positions (vertices)
- N dimensions, N=1, 2, 3, ...
- With or without connectivity information
 - Structured
 - Unstructured
 - Scattered





Grid Structure

Classification

- Geometry
 - Position of vertices in Euclidean space
 - Structured / unstructured
- Topology
 - Cells
 - Connectivity information
 - Neighborhood definition
 - Structured / unstructured

Grid Geometry

Uniform

- implicit relationship between points
- positions can be computed (procedural)



Arecibo Message

- Way of understanding mechanics of raster image representation
- Radio telecope in Puerto Rico
- built in 1964, renovated in 1974
- To celebrate: Frank Drake and Carl Sagan (Cornell University) sent message to M13 in Hercules (25,000 light years away)
- 1679 bits, frequency modulate 2380 MHz



The Message

1679 bits were encoded as 2380MHz plus and minus some frequency

 $1\,1\,1\,0\,0\,0\,0\,1\,1\,1\,0\,0\,0\,0\,0\,1\,1\,0\,1\,1\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,1\,1\,1\,0\,0\,1\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,0\,1\,1\,1\,1\,1\,1\,0\,0\,1\,0\,0$ 01111001111101001111000

This is a **1-D** sequence of bits in time How will an alien understand this list of bits? (will have different symbols than "0" "1") No meta-information!

Understanding the message

- Perhaps some "visual" representation of bits
 - (what is black vs white?)
- Aliens notice 1679 = 23 x 73 (product of two primes)
- Perhaps its not a linear sequence: 2-D array
- •Two ways of sequencing values in 2D array
- •Various ways of laying them out in 2D space
- •Then: have to decipher it!





23 x 73: what was different?

23 x 73: what was different?



73 x 23 . . . 23 x 73: what was different?



73 x 23 compare to:



compare to: http://en.wikipedia.org/wiki/Arecibo_message

23 x 73: what was different?



4 basic pieces of image metadata

- Interpretation of individual values
 - units, scalars, vectors, tensors, measurement frame
- Dimension of array
 - dimension of domain sampled
 - # axes, or # indices for getting a single sample
- Choice of axis ordering (fast-to-slow, or slow-to-fast)
 - Culturally specific
- # samples along each axis
 - "640-by-480 image" or "N-by-M matrix"

Grid Geometry

Structured

- implicit relationship between points
- positions can be computed (procedural)



Grid Geometry

Unstructured

- No underlying structure
- Requires explicit knowledge of every vertex's position: (x₀, y₀, z₀), (x₁, y₁, z₁), ...,



Structured (quadrilateral / hexahedron)

- Implicit connectivity between vertices
- Implicit cell definition



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Unstructured (any cell type)

- Explicit cell definition
 - Types
 - Vertices









Grid Types

		Topology	
		Structured	Unstructured
	Uniform	Image	Unstructured
Geomet	Structured	Rectilinear	Unstructured
	Unstructured	Curvilinear	Unstructured

• Mesh-free (no grid, no connectivity)





Grids (Meshes)

- Meshes combine positional information (geometry) with topological information (connectivity).
- Mesh type can differ substantial depending in the way mesh cells are formed.









(b) Quadratic Triangle







(e) Bi-Quadratic Quad



(d) Quadratic Quad







(f) Quadratic Tetrahedron

(g) Quadratic Pyramid



(h) Quadratic Hexahedron



(i) Bi-Quadratic Hexahedron



(k) Quadratic Linear Wedge



(1) Quadratic Wedge



(j) Tri-Quadratic Hexahedron



(m) Bi-Quadratic Wedge

NONLINEAR CELLS

INTERPOLATION

Mesh Choice Impacts How the Continuous Data is Interpreted

- Two key questions:
 - Sampling, or the choice of where attributes are measured
 - Interpolation, or how to model the attributes in the rest of space



Interpolation

- **Continuous** reconstruction of **discrete** input data $F:\mathbb{R}^n \to \mathbb{R}^n$ $(\mathbf{x}_i, f_i) \to \mathsf{value}$ $\forall i \in \{1, ..., n\}, F(\mathbf{x}_i) = f_i$
- Depends on grid structure (when available)
- Interpolation vs. approximation

Nearest Neighbor Interpolation

- Consider a 1-dimensional, grayscale image I spread horizontally
- What value is I[1.3]?



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• In rectangle

Combination of two consecutive linear interpolation

.

• In rectangle

$$P = P_1 + u(P_2 - P_1) + v(P_4 - P_1) + uv(P_1 - P_2 + P_3 - P_4)$$

Bilinear Interpolation

- Alternate interpretation is a weighted sum of the four pixel values
- Weights defined by the area opposite each corner

http://en.wikipedia.org/wiki/Bilinear_interpolation

Trilinear Interpolation

- In a cuboid (axis parallel)
 - general formula

 $\phi(x,y,z) = axyz + bxy + cxz + dyz + ex + fy + gz + h$

• with local coordinates

But Also...

Nearest Neighbor interpolation

Voronoi diagram

But Also...

- Higher-order interpolation schemes
 - splines, local polynomial fit (interpolation, least sq., ...)
 - smooth reconstruction kernels (on uniform grids)

L17: Isosurfaces REQUIRED READING

Chapter 8

Arrange Spatial Data

For datasets with spatial semantics, the usual choice for arrange is to *use* the given spatial information to guide the layout. In this case, the choices of *express*, *separate*, *order*, and *align* do not apply because the position channel is not available for directly encoding attributes. The two main spatial data types are geometry, where shape information is directly conveyed by spatial elements that do not necessarily have associated attributes, and spatial fields, where attributes are associated with each cell in the field. (See Figure 8.1.) For scalar fields with one attribute at each field cell, the two main visual encoding idiom families are isocontours and direct volume rendering. For both vector and tensor fields, with multiple attributes at each cell, there are four families of encoding idioms: flow glyphs that show local information, geometric approaches that compute derived geometry from a sparse set of seed points, texture approaches that use a dense set of seeds, and feature approaches where data is derived with global computations using information from the entire spatial field.

MARCHING CUBES: A HIGH RESOLUTION 3D SURFACE CONSTRUCTION ALGORITHM

(acm)

William E. Lorensen Harvey E. Cline

General Electric Company Corporate Research and Development Schenectady, New York 12301

Abstract

We present a new algorithm, called *marching cubes*, that creates triangle models of constant density surfaces from 3D medical data. Using a divide-and-conquer approach to generate inter-slice connectivity, we create a case table that defines triangle topology. The algorithm processes the 3D medical data in scan-line order and calculates triangle vertices using linear interpolation. We find the gradient of the original data, normalize it, and use it as a basis for shading the models. The detail in images produced from the generated surface models is the result of maintaining the inter-slice connectivity, surface data, and gradient information present in the original 3D data. Results from computed tomography (CT), magnetic resonance (MR), and single-photon emission computed tomography (SPECT) illustrate the quality and functionality of *marching cubes*. We also discuss improvements that decrease processing time and add solid modeling capabilities.

CR Categories: 3.3, 3.5

Additional Keywords: computer graphics, medical imaging,

acetabular fractures [6], craniofacial abnormalities [17,18], and intracranial structure [13] illustrate 3D's potential for the study of complex bone structures. Applications in radiation therapy [27,11] and surgical planning [4,5,31] show interactive 3D techniques combined with 3D surface images. Cardiac applications include artery visualization [2,16] and nongraphic modeling applications to calculate surface area and volume [21].

Existing 3D algorithms lack detail and sometimes introduce artifacts. We present a new, high-resolution 3D surface construction algorithm that produces models with unprecedented detail. This new algorithm, called *marching cubes*, creates a polygonal representation of constant density surfaces from a 3D array of data. The resulting model can be displayed with conventional graphics-rendering algorithms implemented in software or hardware.

After describing the information flow for 3D medical applications, we describe related work and discuss the drawbacks of that work. Then we describe the algorithm as well as efficiency and functional enhancements, followed by case studies using three different medical imaging techniques to illustrate the new algorithm's carebilities